

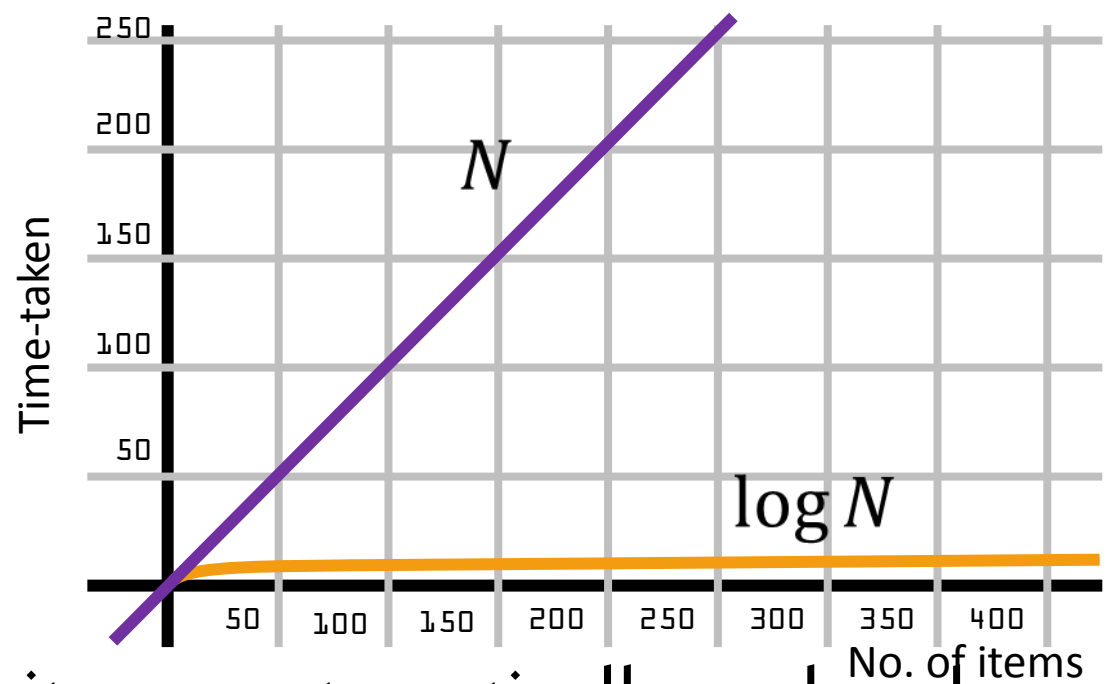
# Sorting algorithms

ESC101: Fundamentals of Computing

Nisheeth

# What is Sorting?

Useful in itself – internet search and recommendation systems



Sorting is the process of arranging items systematically, ordered by some criterion

Search within  $n$  unsorted elements can take as much as  $O(n)$  operations

Makes searching very fast – can search within  $n$  sorted elements in just  $O(\log n)$  operations using **Binary Search**

# Sorting Algorithms

Bubble Sort



# Bubble Sort

- Consider an array (5 1 4 2 8). Goal: Sort it in ascending order
- Idea: Repeatedly **swap the adjacent elements** if they are in **wrong order**

## First Pass

( **5** 1 4 2 8 )  $\rightarrow$  ( 1 **5** 4 2 8 )  
( 1 **5** 4 2 8 )  $\rightarrow$  ( 1 4 **5** 2 8 )  
( 1 4 **5** 2 8 )  $\rightarrow$  ( 1 4 2 **5** 8 )  
( 1 4 2 **5** 8 )  $\rightarrow$  ( 1 4 2 5 **8** )

## Second Pass

( **1** 4 2 5 8 )  $\rightarrow$  ( 1 **4** 2 5 8 )  
( 1 **4** 2 5 8 )  $\rightarrow$  ( 1 2 **4** 5 8 )  
( 1 2 **4** 5 8 )  $\rightarrow$  ( 1 2 4 **5** 8 )  
( 1 2 4 **5** 8 )  $\rightarrow$  ( 1 2 4 5 **8** )

## Third Pass

( **1** 2 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 **2** 4 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 **4** 5 8 )  $\rightarrow$  ( 1 2 4 5 8 )  
( 1 2 4 **5** 8 )  $\rightarrow$  ( 1 2 4 5 8 )



No swaps in this pass,  
hence done! 😊



# Bubble Sort

```
// A function to implement bubble sort
void bubbleSort(int arr[], int n)
{
    int i, j;
    for (i = 0; i < n-1; i++)

        // Last i elements are already in place
        for (j = 0; j < n-i-1; j++)
            if (arr[j] > arr[j+1])
                swap(&arr[j], &arr[j+1]);
}
```

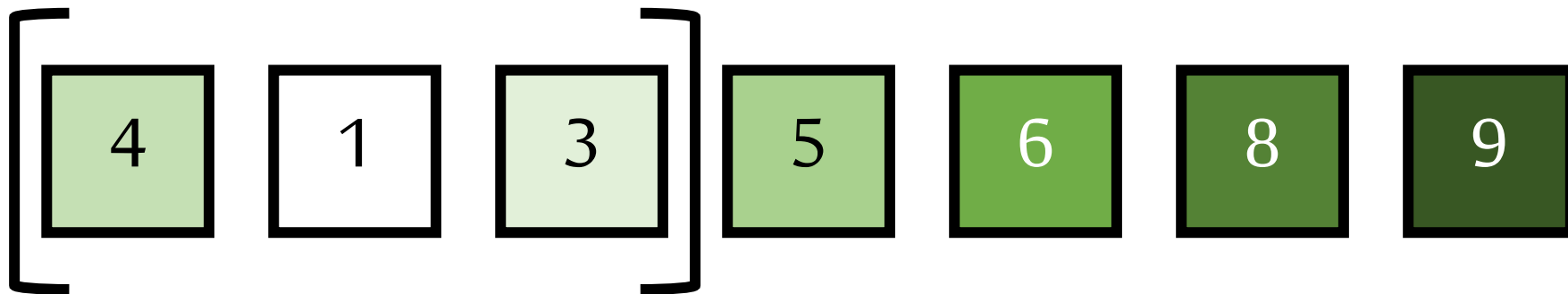
# Sorting Algorithms

## Selection Sort



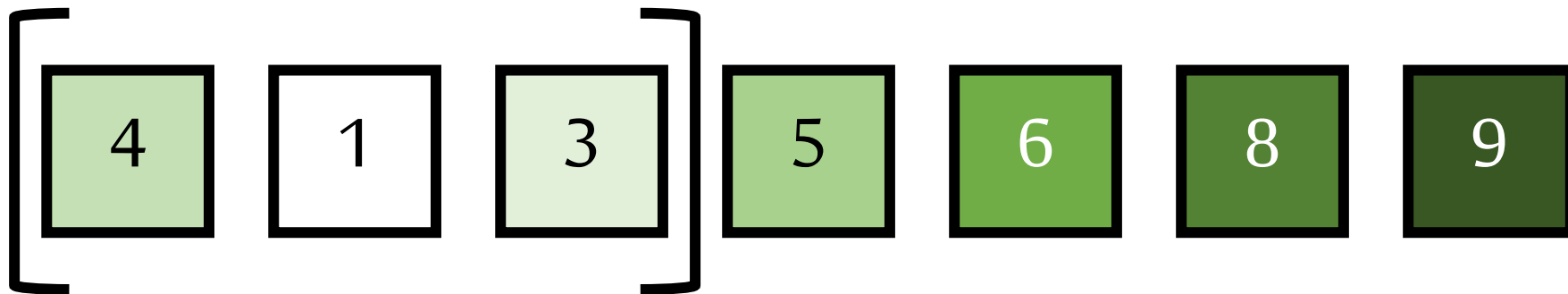
# Selection Sort

- Another very simple sorting algorithm
- Like binary search, maintains *active range*  $a[0:R]$  with  $0 \leq R < N$ 
  - Initially the active range is entire array i.e.  $R = N - 1$
- We will ensure two things
  - At all points of time, the *non-active portion* will be sorted in ascending order i.e. for all  $R \leq i < j$  we will ensure  $a[i] \leq a[j]$
  - The non-active elements will never be smaller than the elements in the active range i.e. if  $i \leq R < j$  then  $a[i] \leq a[j]$
- Active region will shrink by one element at each step (details in next class)



# Selection Sort

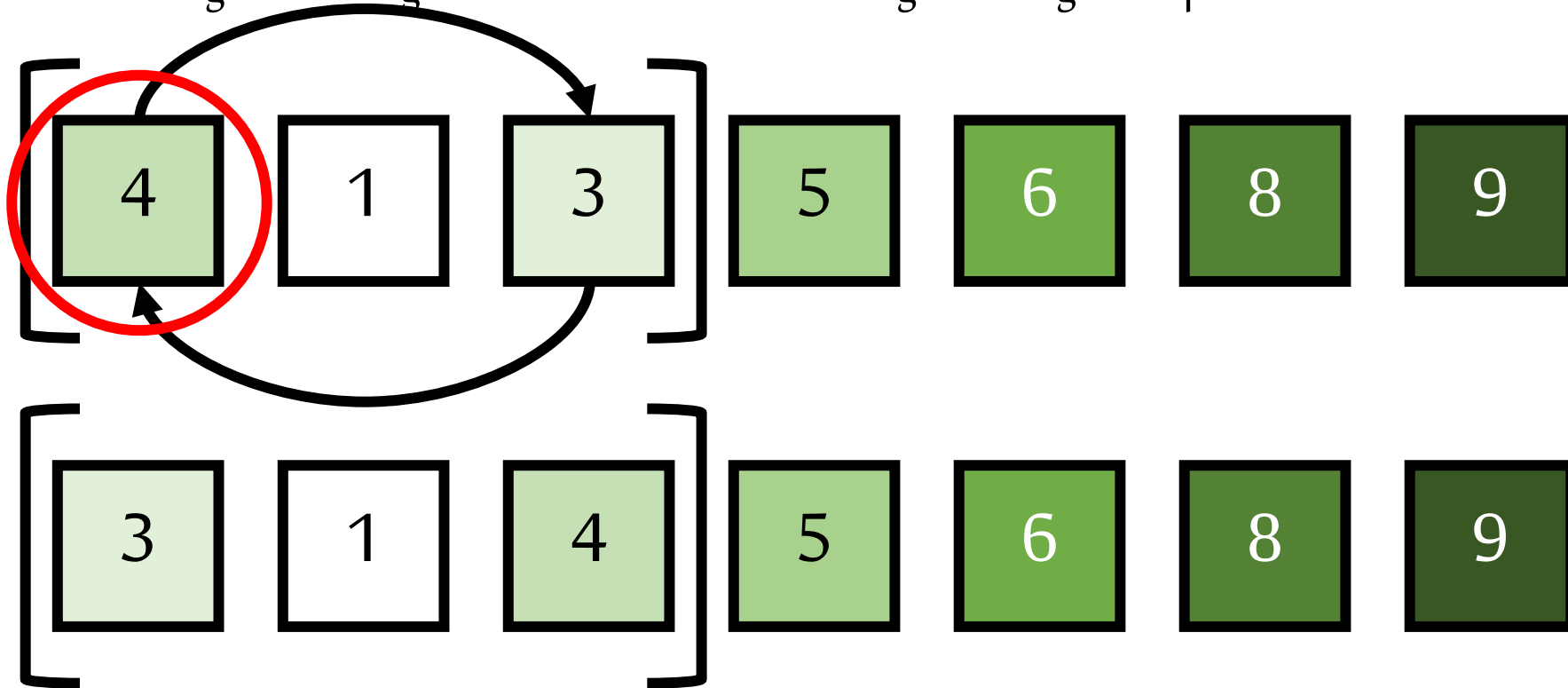
- Already saw Bubble Sort. Selection sort is another very simple sorting algo
- Like binary search, maintains **active range**  $a[0:R]$  with  $0 \leq R < N$ 
  - Initially the active range is entire array i.e.  $R = N - 1$
- We will ensure two things
  - At all points of time, the **non-active portion** will be sorted in ascending order i.e. for all  $R \leq i < j$  we will ensure  $a[i] \leq a[j]$
  - The non-active elements will never be smaller than the elements in the active range i.e. if  $i \leq R < j$  then  $a[i] \leq a[j]$
- The active region will **shrink by one element** at each step





# Selection Sort

- Once an element goes to non-active region, we never touch it again 😊
- To maintain our conditions and still shrink the active region
  - We find the largest element in the active region
  - Bring it to the right-most end of the active region using a swap



Verify that our conditions  
still hold

# Selection Sort

Exercise: write a recursive version

Exercise: convert this to proper C code

## SELECTION SORT

1. Given: Array  $a$  with  $N$  elements
2. For  $R = N - 1; R > 0; R - -$  // Initial active range is full array
  1.  $i \leftarrow \text{FINDMAX}(a, 0, R)$  // Location of largest element in  $a[0, R]$
  2.  $\text{SWAP}(a, i, R)$  // Bring largest element to the end

## SWAP

1. Given: Array  $a$ , location  $i, j$
2. Let  $tmp \leftarrow a[i]$
3. Let  $a[i] \leftarrow a[j]$
4. Let  $a[j] \leftarrow tmp$

## FINDMAX

1. Given: Array  $a$ , locations  $i, j$
2. Let  $k \leftarrow i, \text{max} = a[k]$
3. For  $l = i; i \leq j; l + +$ 
  1. If  $a[l] > \text{max}, \text{max} = a[l], k = l$
4. Return  $k$

# Time Complexity

- Let  $T(N)$  be the time taken for selection sort to sort  $N$  elements
- Let  $M(N)$  be the time taken to find location of max of  $N$  elements
- At any time step when active region is  $[0: R]$ , we do two things
  - Find the largest element within the active region – takes time  $M(R + 1)$
  - Swap the largest element with the element at  $a[R]$  - takes time  $c$  (const)
- Thus, we have  $T(N) \leq M(N) + c + T(N - 1)$
- It is easy to show that  $M(N) \leq d \cdot N$  for all  $N$  for some constant  $d$
- Exercise: expand the recurrence as before and show that

$$T(N) \leq \mathcal{O}(N^2)$$

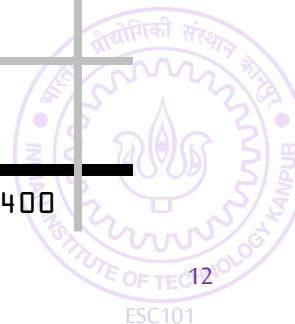
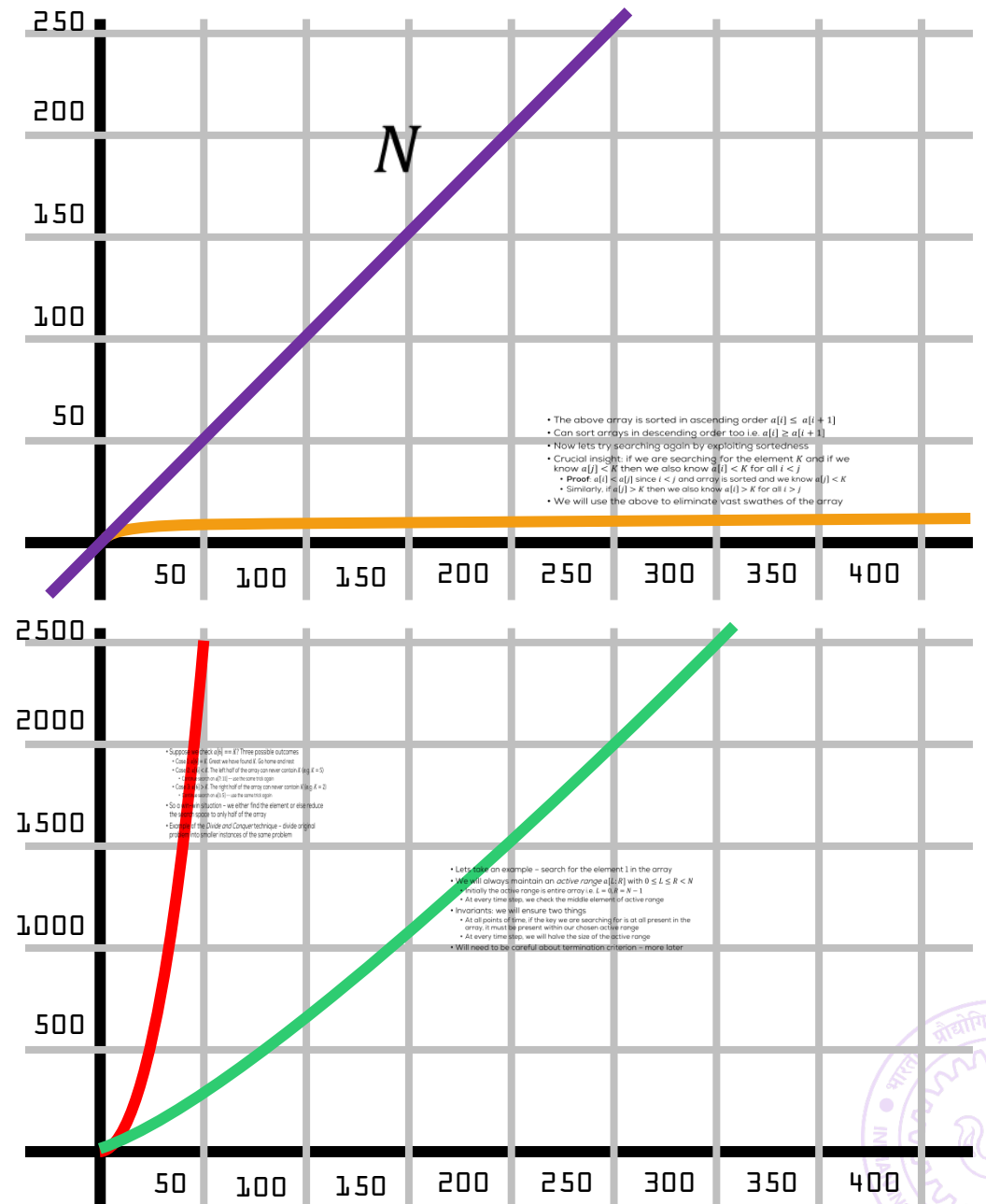
Assume  $T(1) \leq c$

- Notice that selection sort (also bubble sort) **doesn't need any extra memory** (except a few tmp variables to store one integer each) – ***in-place sorting***



# Summary so far..

- Applications of sorting: ranking, recommendation, internet search
- Brute force search  $\mathcal{O}(N)$
- Fast searches on sorted arrays: binary search  $\mathcal{O}(\log N)$
- Bubble sort  $\mathcal{O}(N^2)$
- Selection sort  $\mathcal{O}(N^2)$
- Next: fast sorting  $\mathcal{O}(N \log N)$ 
  - Merge Sort
  - Quick Sort



# Partition based Sorting Techniques

- Let  $T(N)$  be the time taken for selection sort to sort  $N$  elements
- Let  $M(N)$  be the time taken to find location of max of  $N$  elements
- For selection sort, we saw that  $T(N) \leq M(N) + c + T(N - 1)$
- Active region **shrank too slowly** which gave us  $T(N) \leq \mathcal{O}(N^2)$
- Selection sort (also bubble sort) is quite expensive (imagine  $\mathcal{O}(N^2)$  time complexity for  $N = 1,000,000$  items 😞) – much better can be done
- Will study two sorting algorithms based on **divide and conquer** technique
- Both algorithms will split an array of  $N$  elements into two arrays, sort each smaller array and then do some clean up operations
  - Merge Sort: popular for sorting large scale databases
  - Quick Sort: extremely popular in general (see `qsort()` in `stdlib.h`)

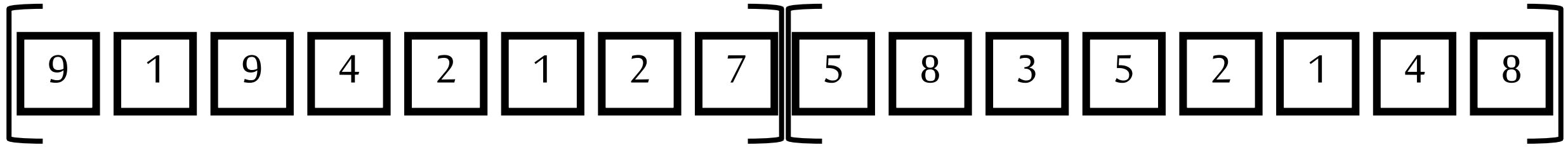


# Sorting Algorithms

## Merge Sort

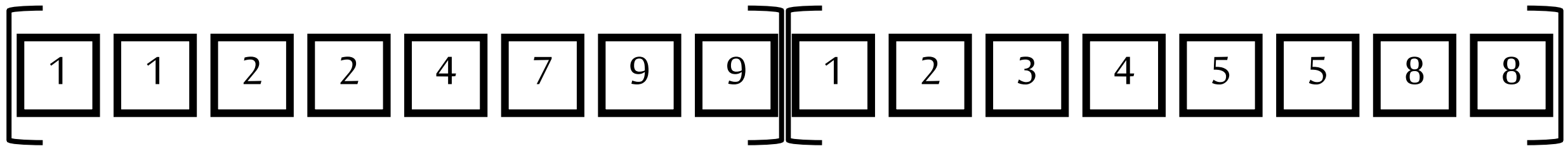


# Merge Sort

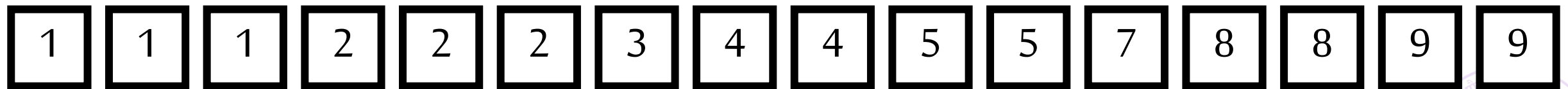


? Merge Sort

? Merge Sort



? Merge



Trick: Merging two sorted arrays is very easy!



# Merge Sort

Why didn't we split as  $[0:N - 2], [N - 1:N - 1]$  ?  
No need to find middle element. Also, would have made one of the mergesort calls so simple!

- B**
1. Given: Sorted array  $a$  with  $N$  elements, key to search  $K$
  2. Let  $L \leftarrow 0$  and  $R \leftarrow N - 1$  // Initial active range is full array
  3. While  $L \leq R$ 
    1. Let  $M \leftarrow \text{ceil}((L + R)/2)$
    2. If  $a[M] == K$ , return  $M$  // If found, return index
    3. If  $a[M] > K$ , set  $R \leftarrow M - 1$  // If current element is greater than key, search left
    4. If  $a[M] < K$ , set  $L \leftarrow M + 1$  // If current element is less than key, search right
  4. Return  $-1$  // If not found, return -1

A sort algorithm is called *in-place* if it does not use extra memory e.g. extra arrays, to sort the given array

- An effort to quantify the *speed* of algorithms in a manner that is independent of the computer on which they are executed
- Arguably binary search seems "faster" than brute force search
- We saw that in the worse case, brute force search on an unsorted array must check all  $N$  elements before answering
- Can binary search on sorted arrays also be forced to do so?
- Let  $T(N)$  denote the time taken by binary search to search for a key in a sorted array with  $N$  elements
- We know that at every iteration of the while loop, binary search either discovers the element being searched or else reduces the length of the active range by a factor of 2



# Time Complexity

If we had split as  $[0:N-2], [N-1:N-1]$  then  $T(N) \leq T(N-1) + T(1) + M(N)$  would have given us  $T(N) = \mathcal{O}(N^2)$  (divide properly to rule powerfully 😊)

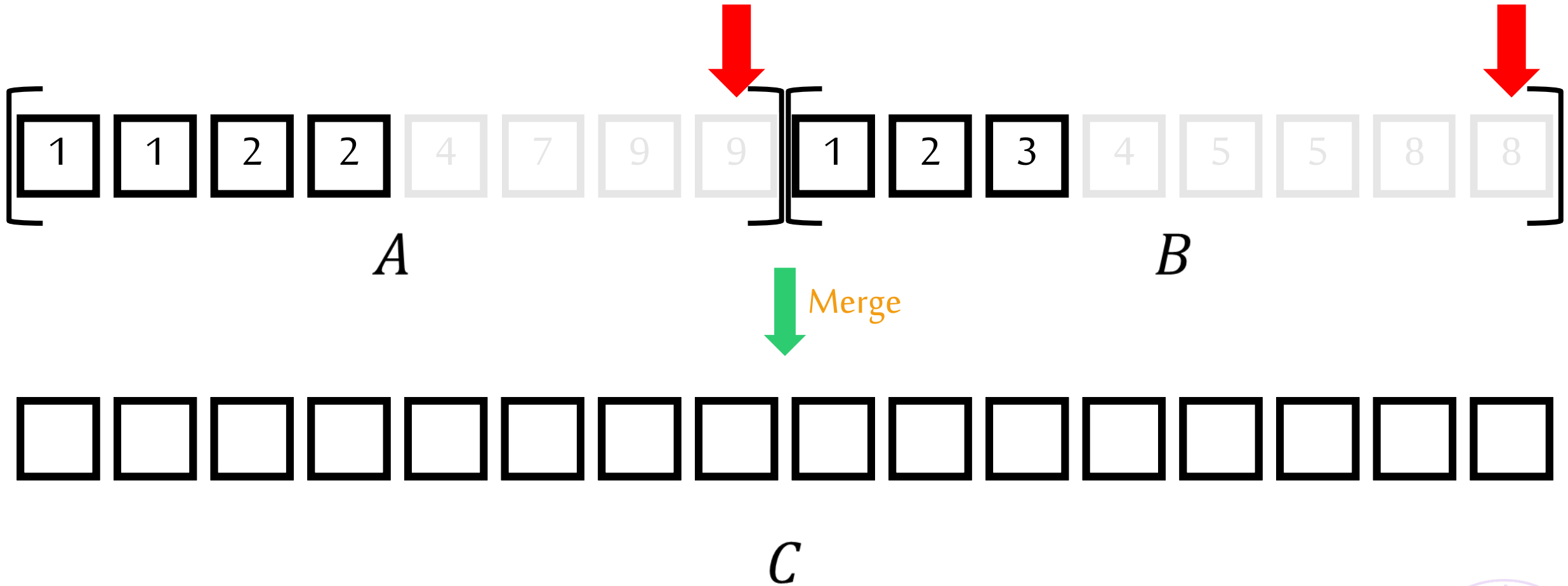
- Let  $T(N)$  be the time taken for sorting  $N$  elements
- Let  $M(N)$  be time merging two sorted arrays with total  $N$  elements
- Thus, we have  $T(N) \leq 2 \cdot T(N/2) + M(N) + d$  ( $d$ : time to find middle index)
- We will show next that we can do  $M(N) \leq c \cdot N$  time
- This recurrence is a bit harder to solve but we can still try
$$T(N/2) \leq 2 \cdot T(N/4) + c \cdot N/2 + d$$
$$T(N) \leq 4 \cdot T(N/4) + 2c \cdot N + (1 + 2) \cdot d$$
$$T(N) \leq 2^k \cdot T(N/2^k) + kc \cdot N + 2^k \cdot d$$
- Set  $k = \text{ceil}(\log N)$  and use  $T(1) \leq c$  to get  $T(N) \leq \mathcal{O}(N \log N)$
- The version of merging we will show uses extra  $\mathcal{O}(N)$  memory. Can you develop a version that uses only 2-3 extra integer variables i.e. an *in-place* version of merge sort?

# The Merge Operation

- Given 2 arrays int  $a[M], b[N]$ ; both sorted in ascending order
- Want a combined array int  $c[M + N]$ ; sorted in ascending order
- Will maintain *active ranges* for both arrays  $a[0: R_1]$  and  $b[0: R_2]$  with  $0 \leq R_1 < M$  and  $0 \leq R_2 < N$ 
  - Initially the active ranges are the entire arrays i.e.  $R_1 = M - 1, R_2 = N - 1$
- At all points of time, we will ensure that elements in the non-active regions of the arrays would have been inserted into  $c$  at their proper locations
- At least one active region will shrink by one element at each step
- Trick: the largest element of  $c$  can be found in  $\mathcal{O}(1)$  time since the arrays  $a, b$  are sorted. If unsorted it would have taken  $\mathcal{O}(M + N)$



# The Merge Operation



$\theta$  is larger: B wins again

# The Merge Operation

- Given a (non-sorted) array `int a[N]`; count the number of swaps  
A swap is a pair  $0 \leq i \neq j < N$  such that  $i < j$  but  $a[i] > a[j]$ 
  - This problem is related to a ranking metric known as *area under the ROC curve*. Check it out if interested
- We have two arrays of  $N$  numbers `int P[N], F[N]`; containing 12<sup>th</sup> marks of  $N$  students each who cleared and did not clear JEE
  - Find out the number of students who did not clear JEE but had 12<sup>th</sup> std marks more than at least 50% of students who did clear JEE
- Solve these problems faster than  $O(N^2)$  time (Hint may involve sorting). Assume you have a routine that can sort  $N$  elements in  $O(N \log N)$  time – will see such methods soon.



# Sorting Algorithms

## Quick Sort



# Quick Sort

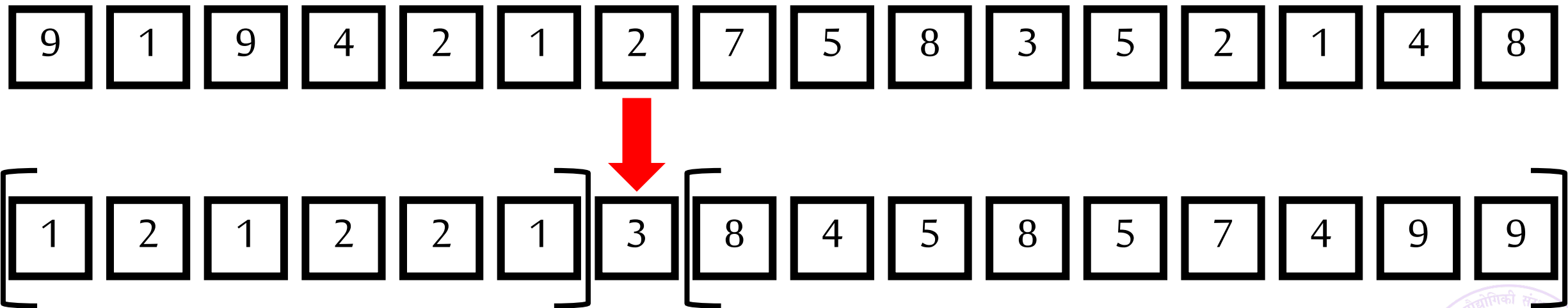
- Very popular sorting algorithm – try this before anything else
- $\mathcal{O}(N \log N)$  time complexity but in practice faster than merge sort
- Like selection sort, merge sort lazily divides the array into two equal halves, sorts the halves recursively and then spends time merging them
- Quick sort is **more careful in splitting** the array so that **no need for merging** once the subarrays are sorted!
- Based on a cool trick known as *partitioning*
- Analysis of quick sort is much more advanced – in worst case quicksort takes  $\mathcal{O}(N^2)$  time but this happens very very rarely.
- On average quicksort enjoys  $\mathcal{O}(N \log N)$  time complexity





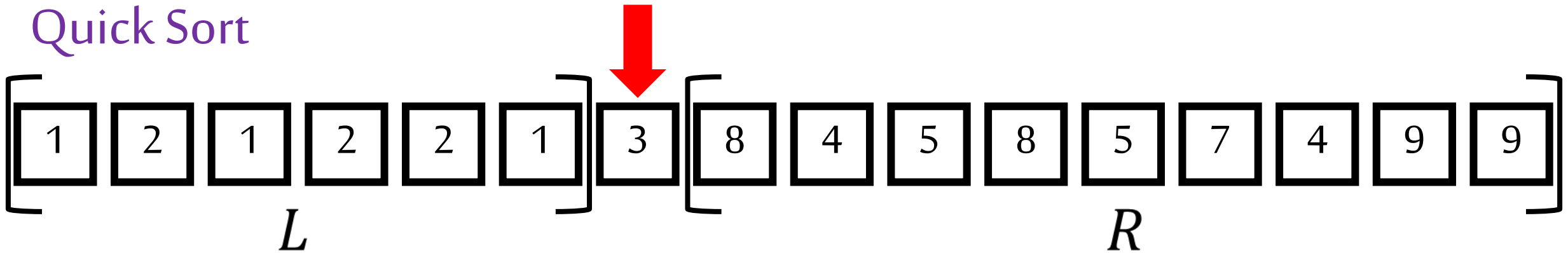
# The Partition Technique

- Given array  $\text{int } a[N]$  and any element of the array  $p$  (called pivot)
- Create a new array  $\text{int } b[N]$  which is arranged as follows  
[elements of  $a \leq p$ ,  $p$ , elements of  $a \geq p$ ]



- Notice that left and right halves are not sorted yet! 😊
- Also, the two halves are not balanced (of same size) either 😊

# Quick Sort



- Notice that even though the subarrays  $L, R$  not sorted, every element of  $L$  is smaller than or equal to every element of  $R$
- This means that if we sort  $L, R$  recursively, no need to merge 😊
- Key to quicksort's success – partition and recursively sort!
- Will discuss a partition algorithm that ensures a stricter condition  
[elements of  $a < p$ , all instances of  $p$ , elements of  $a > p$ ]
- However, our algorithm will use extra memory
- Time complexity analysis of quicksort beyond scope of ESC101



# Quick Sort

## QUICKSORT

1. Given: Array  $a$  with  $N$  elements.  $i$  is the new location of the pivot element
2. If  $N < 2$  return  $a$  // or singleton array is sorted
3. Let  $p \leftarrow \text{CHOOSEPIVOT}(a)$  // Choose a pivot value
4. Let  $(b, i) \leftarrow \text{PARTITION}(a, p)$  // Partition along chosen pivot
5.  $\text{QUICKSORT}(b[0:i - 1])$  // Sort the left half
6.  $\text{QUICKSORT}(b[i + 1, N - 1])$  // Sort the right half
7. Return  $b$

Common choices for pivot value

- $a[0]$  or  $a[N - 1]$  i.e. end elements
- $a[i]$  for  $i \sim \text{random}(N)$  i.e. a random element
- $\text{MEDIAN}(a)$  i.e. median element of the array

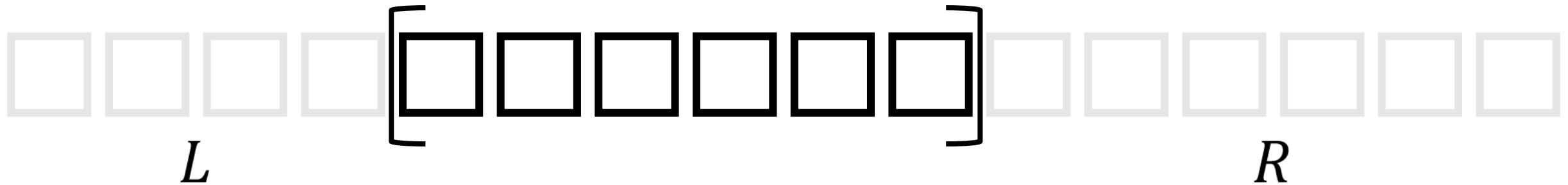
Most popular, inexpensive

Also common, inexpensive

Ensures balanced partition but expensive

# The Partition Procedure

- The partition procedure maintains an interesting structure of one active region sandwiched between two inactive regions 😊



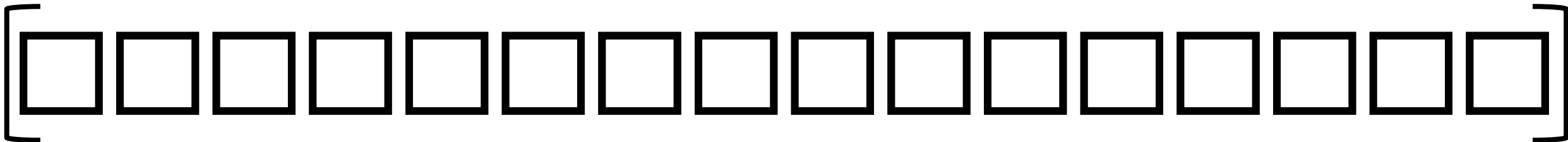
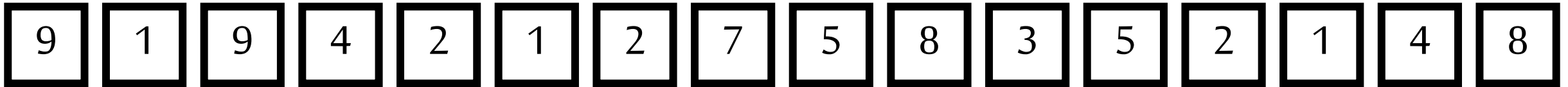
- Elements in the left inactive region are strictly less than the pivot, those in right invariant region strictly larger than pivot
- What about element(s) equal to the pivot – need to be careful
- We will see a visualization of the partition procedure in action
- Note: these regions will be maintained on a separate array and not the original array – we will only take a simple left-to-right pass on the original array

We are sure now that any blank spaces left must be occurrences of pivot 4 that we omitted earlier

All occurrences of the pivot element are done with non-pivot elements

PIVOT =

4



*L*

*R*

8 < 4 i.e. belongs to *R*

Can't insert 4 now as there are still elements of *L/R* left to be processed. If we insert 4 now, we may violate our conditions later

# The Partition Procedure

## PARTITION

1. Given: Array  $a$  with  $N$  elements, pivot  $p$
2. Let  $\text{int } b[N], L \leftarrow 0, R \leftarrow N$  // Initial boundaries
3. For  $i = 0; i < N; i++$ 
  1. If  $a[i] < p$ , let  $b[L] \leftarrow a[i]$ , and  $L++$  // We found a left element
  2. If  $a[i] > p$ , let  $b[R] \leftarrow a[i]$ , and  $R--$  // We found a right element
4. For  $i = L; i \leq R; i++$ 
  1. Let  $b[i] \leftarrow p$  // Fill up the gap
5. Return  $(b, R)$

Verify that after the first loop has ended, we must have  $L < R$  i.e. some space left for pivot

In fact, the entire range  $b[L:R]$  is filled with the pivot element

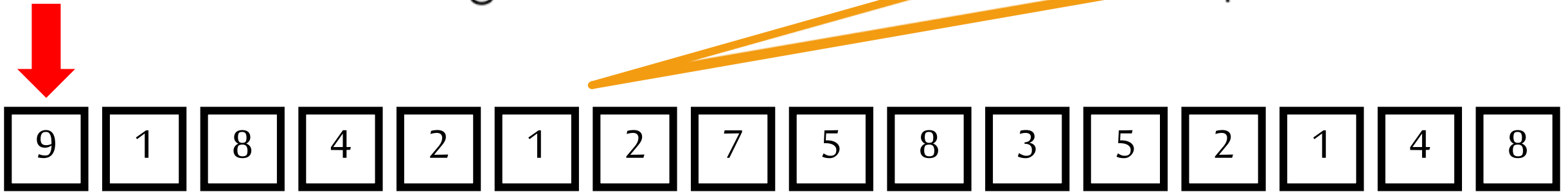
$R$  has to be (one of) the new location(s) of the pivot element

Hint: the in-place algorithm uses an identical notion of inactive regions but swaps elements at the boundaries of the regions which are wrongly placed

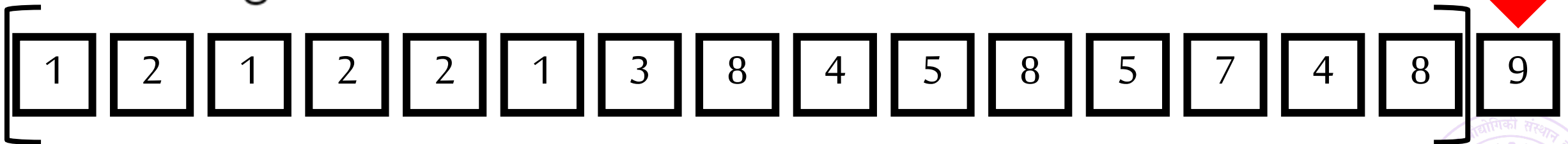
Explore/invent yourself an *in-place* partitioning algorithm

## Choice of Pivot

- Most crucial step in quicksort –
- Suppose we are so unlucky that we choose the smallest or the largest element as a pivot. Ironically, if the array is already sorted and we use end elements as pivots, then quicksort takes  $\mathcal{O}(N^2)$  time ☹️



- Choosing an element close to the median is most beneficial



- Quicksort becomes selection sort i.e.  $\mathcal{O}(N^2)$  time ☹️

## Next class and next week..

- Wrap up the discussion on sorting
- Hashing: a very efficient method for search
- File handling in C
- Solving numerical problems using programming
- Future directions and wrapping up the course

